

Abstract

Assimilation of precipitation has been generally unsuccessful because: a) errors associated with precipitation are not Gaussian, and this violates the basic assumption of current data assimilation methods, and b) forcing the model to rain where observed by moistening or drying the atmosphere, as done in assimilation of precipitation in the past, does not change the main dynamical variable (potential vorticity) and thus is forgotten as soon as the forcing of the “assimilation of moisture” ceases. Lien *et al.* (2013, 2016a, 2016b) and later Kotsuki *et al.* (2017) developed a methodology that successfully addressed these problems: They performed a transformation of observed and modeled precipitation into Gaussian variables (anamorphosis) which made the errors also much more Gaussian. They also used ensemble-based data assimilation approaches that are more naturally able to assimilate precipitation and directly influence the model potential vorticity, since the ensemble members that precipitate closely as observed, receive higher weights in the ensemble analysis, and therefore “donate” their more correct dynamics to the analysis mean, thus creating an analysis mean that is closer to having the right dynamics. As a result, both Lien *et al.* (2016) and Kotsuki *et al.* (2017) showed that assimilation of remotely sensed precipitation (NASA TMPA, and JAXA GSMaP) improves the forecast skill for over 5 days.

The goal of this study is to investigate whether we can improve the Tropical Cyclone (TC) predictions by precipitation assimilation with the Gaussian Transformation. We implement the precipitation assimilation module into the Local Ensemble Transform Kalman Filter (LETKF) assimilation system for the mesoscale model SCALE-RM developed at the RIKEN Advanced Institute for Computational Science in Japan.

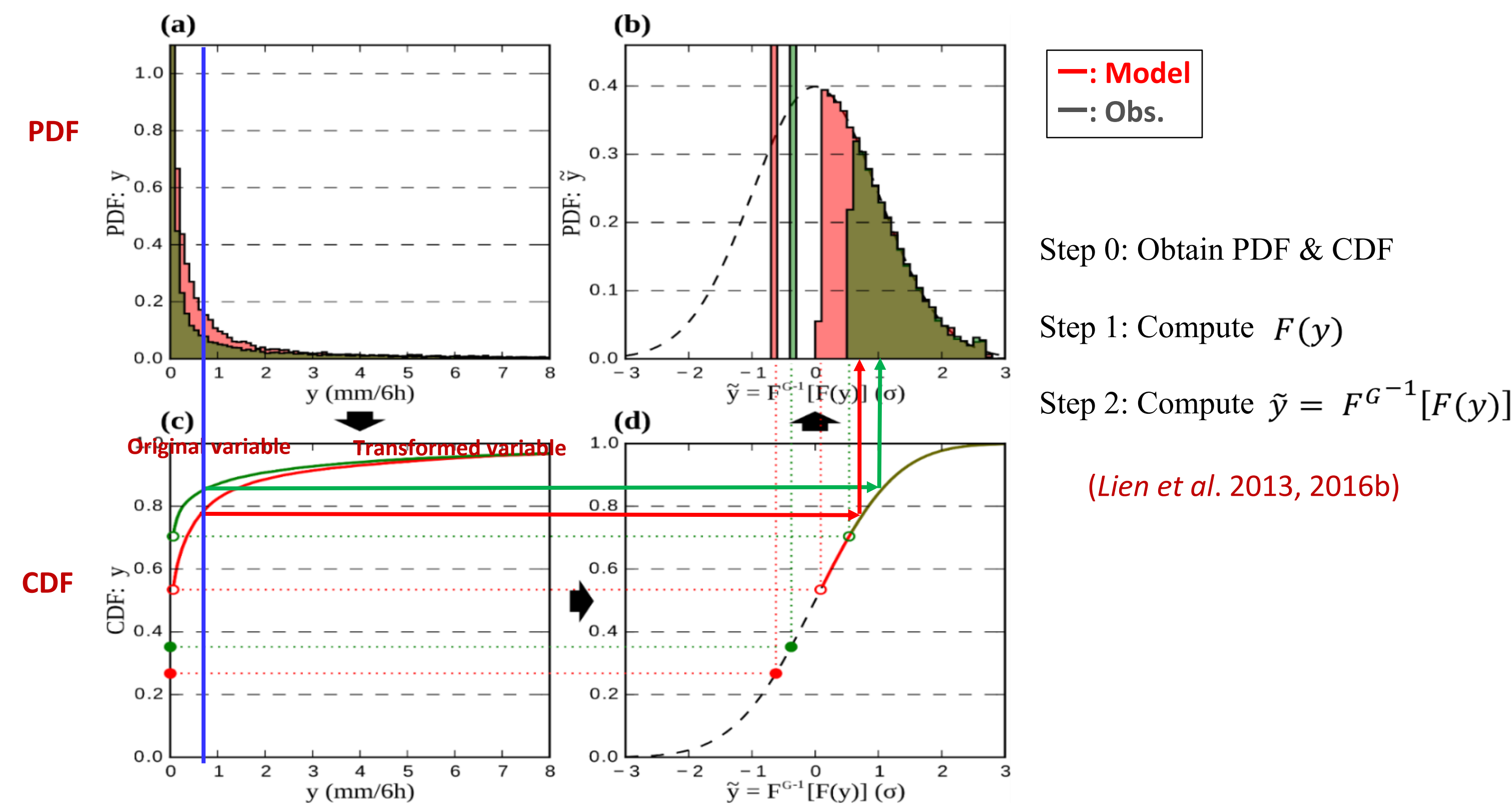
We conducted assimilation experiments for 4 typhoon cases in 2015 by assimilating additional Level-3 JAXA GSMaP retrievals. Our results show that with additional assimilation of JAXA GSMaP retrievals,

- (1) The minimum SLP analysis is closer to the value of the JMA best track datasets;
- (2) The hydrometeor distribution analysis is closer the GMI Level-2 retrievals;
- (3) Both the 3-day track and intensity forecasts are improved.

Gaussian Transformation

$$F^G(\tilde{y}) = F(y) \Rightarrow \tilde{y} = F^{G^{-1}}[F(y)]$$

y : original variable (mm/6hr) \tilde{y} : Transformed variable (sigma)
 $F(\cdot)$: CDF of original variable $F^G(\cdot)$: CDF of Gaussian distribution



Overview of Four Typhoon Cases in 2015

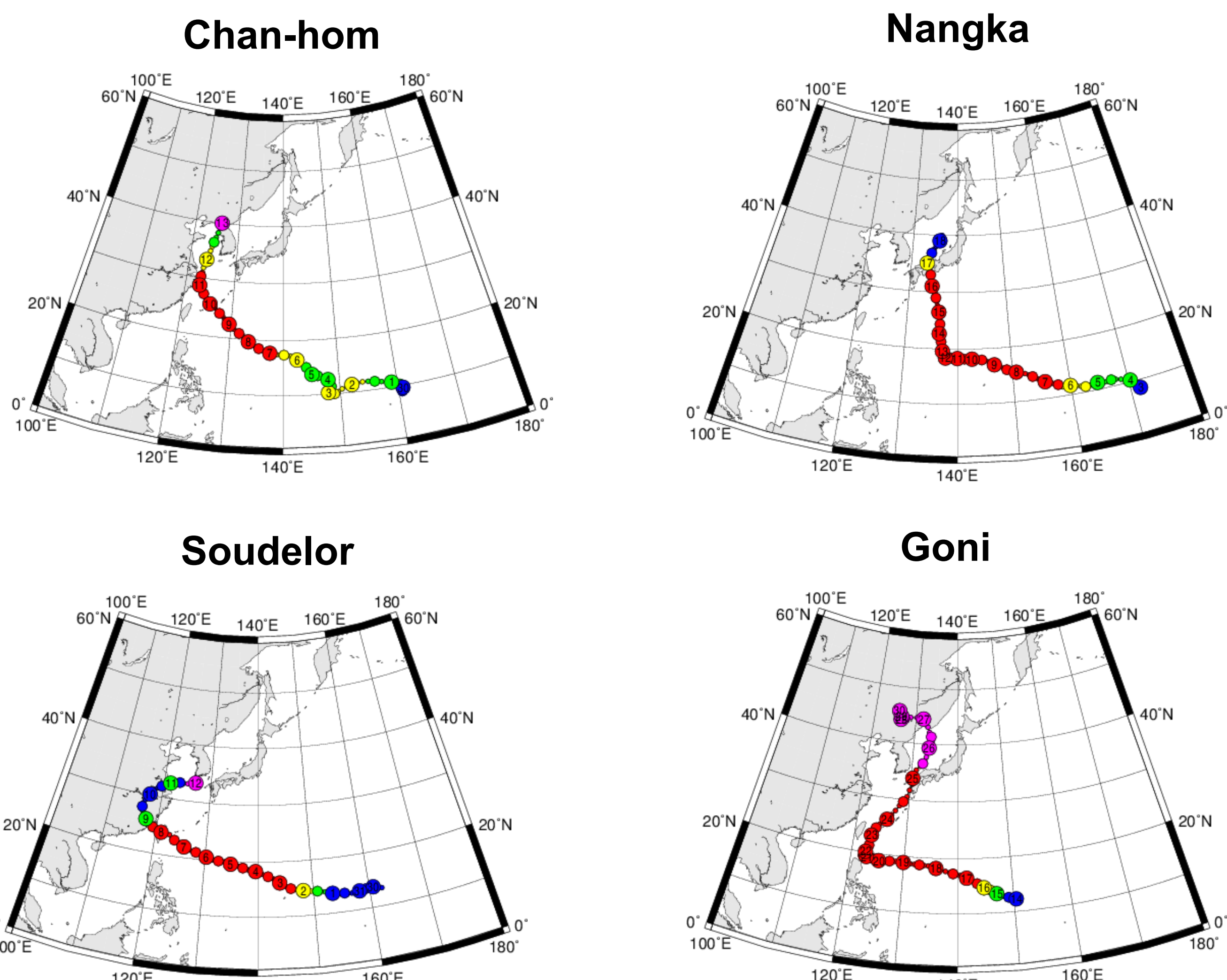


Figure 1: Best track of Typhoon Chan-hom, Nangka, Soudelor, and Goni in 2015. Figures from <http://agora.ex.nii.ac.jp>

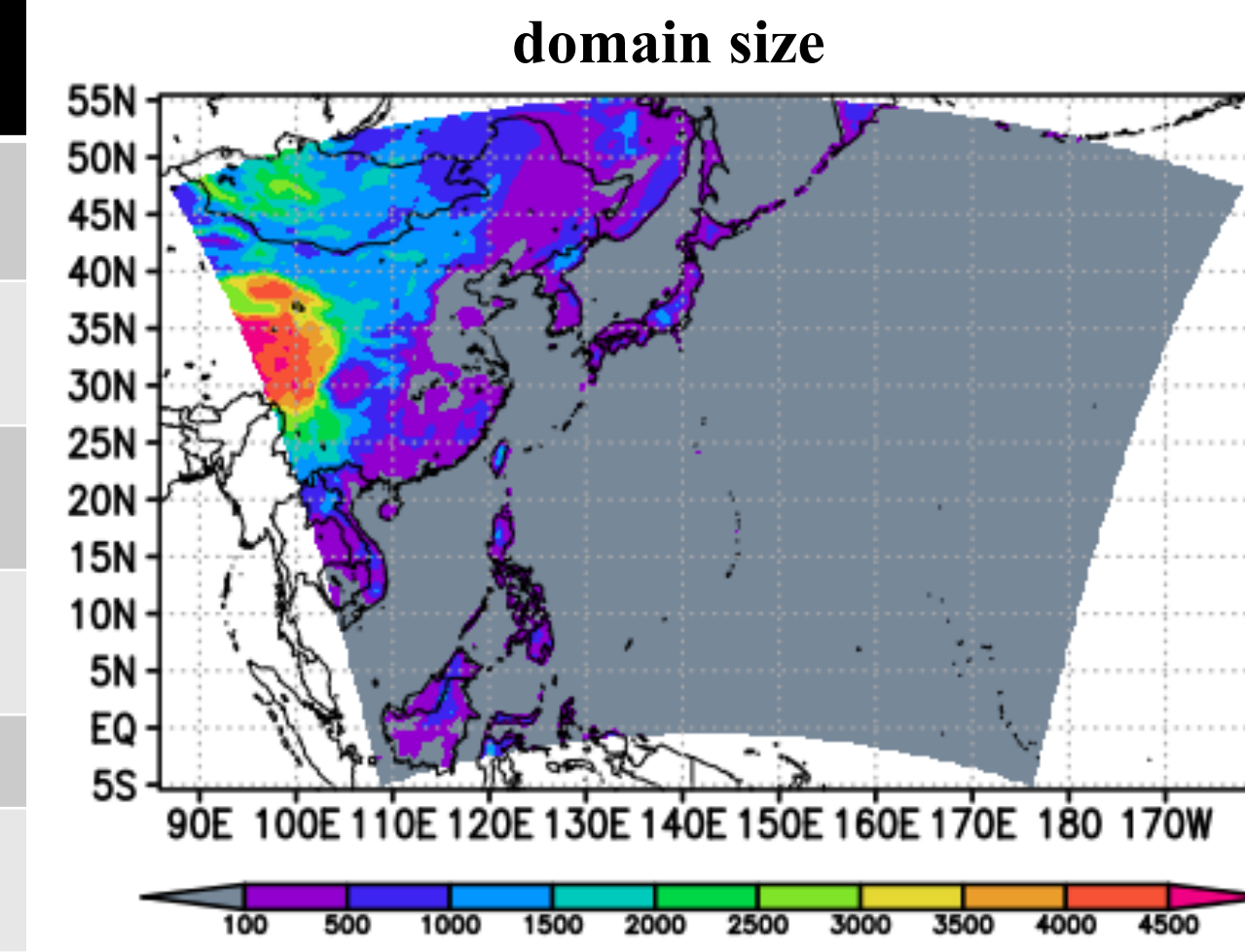
Reference

- GY Lien, E Kalnay, T Miyoshi, 2013: Effective assimilation of global precipitation: Simulation experiments. Tellus A 65, 19915.
- GY Lien, T Miyoshi, E Kalnay, 2016: Assimilation of TRMM multisatellite precipitation analysis with a low-resolution NCEP global forecast system, Monthly Weather Review.
- GY Lien, E Kalnay, T Miyoshi, GJ Huffman, 2016: Statistical properties of global precipitation in the NCEP GFS model and TMPA observations for data assimilation. Monthly Weather Review
- S Kotsuki, T Miyoshi, K Terasaki, GY Lien, E Kalnay, 2017: Assimilating the Global Satellite Mapping of Precipitation Data with the Nonhydrostatic Icosahedral Atmospheric Model NICAM. Journal of Geophysical Research

Model Settings and Experiment Design

SCALE Model (Nishizawa *et al.* 2015; Sato *et al.* 2015)

SCALE Settings	D01
Resolution	36km
Domain size	240×180 grids, 36 levels (~28km)
Boundary Layer	MYNN
Microphysics	TOMITA08 (6 species)
Radiation	MSTRNX
Cumulus Para.	Kain-Fritsch



SCALE-LETKF (Lien *et al.* 2017)

LETKF settings	CTRL	GSMaP
Ensemble size		100
Inflation		Multiplicative inflation, RTPP
BC		GFS FNL analysis every 6 hours
V-Loc.		0.3log(pressure)
H-Loc.		400km (PrepBUFR)/250km (GSMaP)
Obs assimilated.	PrepBUFR	PrepBUFR + GSMaP
Threshold for precipitation		$\geq 0.001\text{mm (6hr)}^{-1}$
Loc. level of rain obs.		850mb
QC		$\geq 75/100$ (75%) have rain.

- Spin-up: 4-days before the formal experiment
- Precipitation CDFs: ~5-day precipitation distribution before spin-up.

Single-observation Experiment

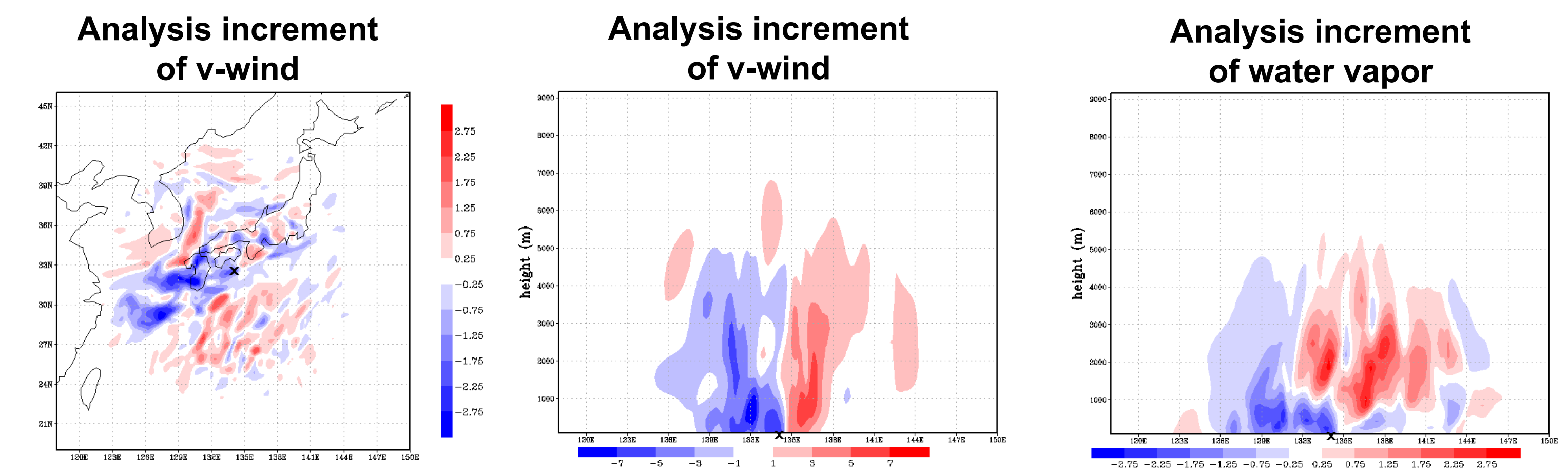
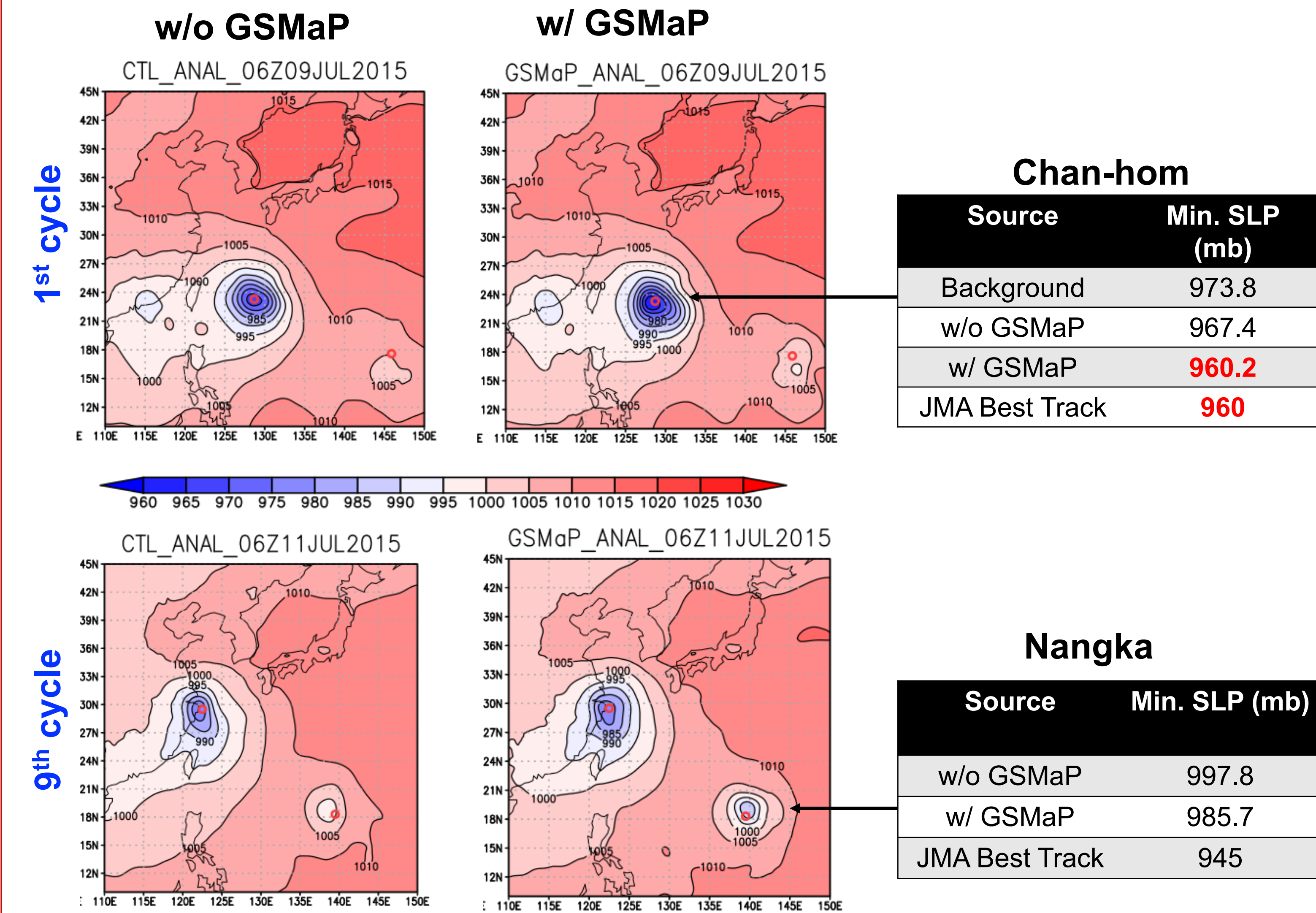
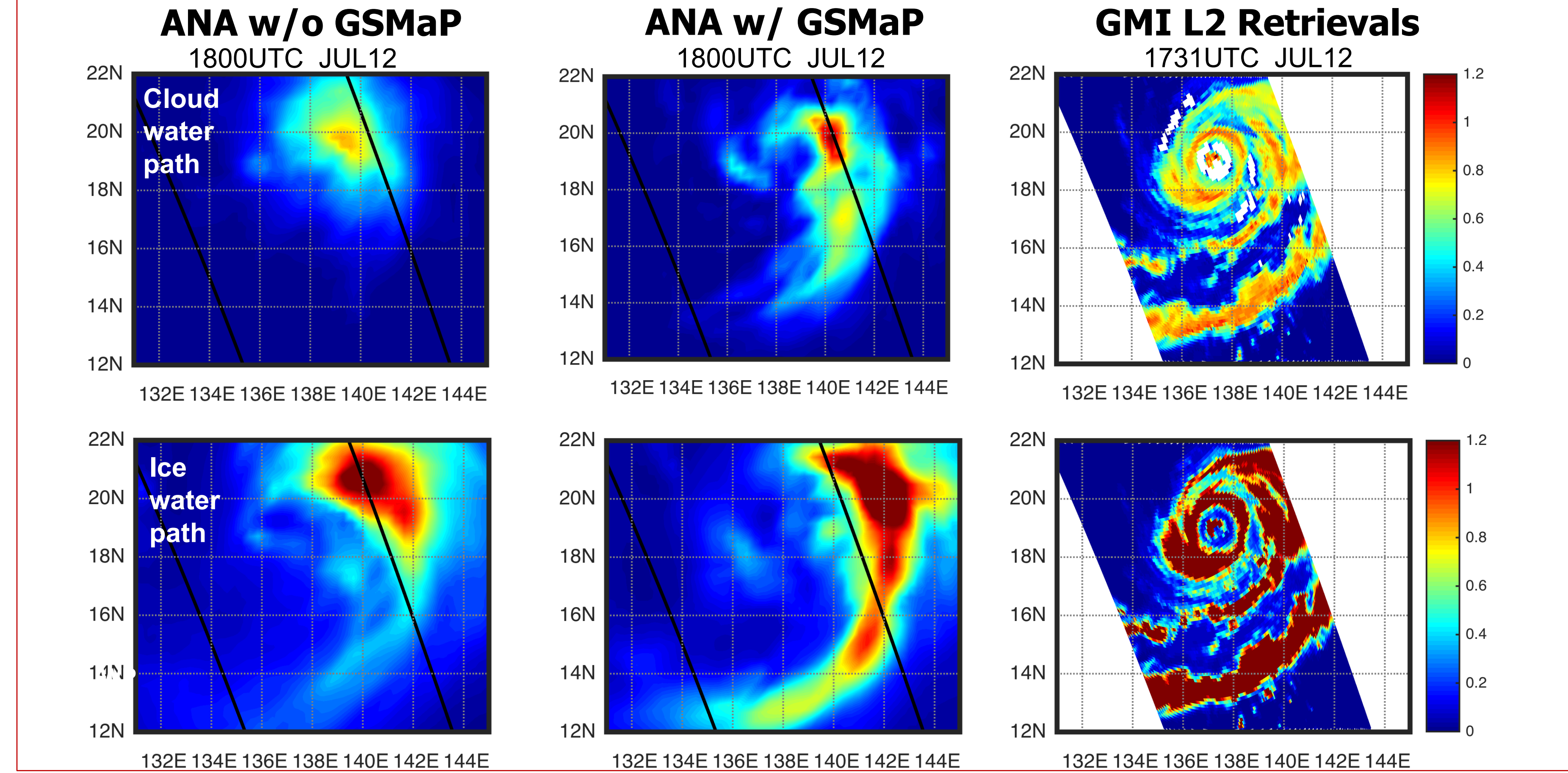


Figure 2: The analysis increment (x^a-x^b) map of the v-wind at the surface (left), and the cross section of the v-wind (middle) and the water vapor (right) along the latitude where the observation resides. The location of the single GSMaP observation is marked by a black cross.

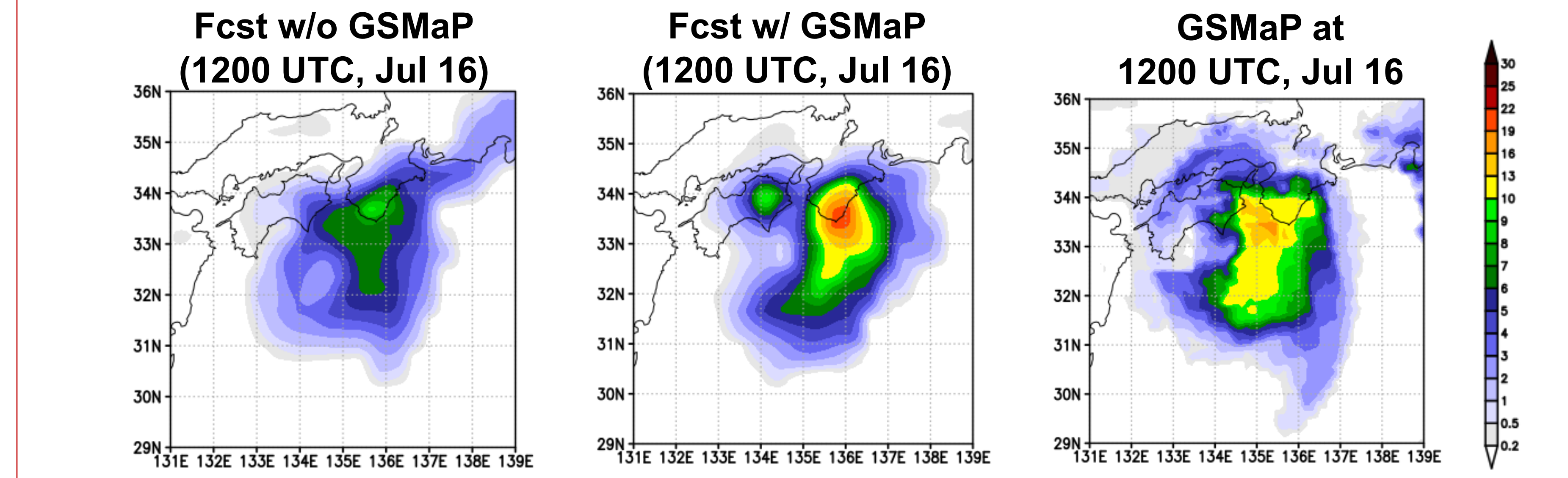
Improved SLP analysis



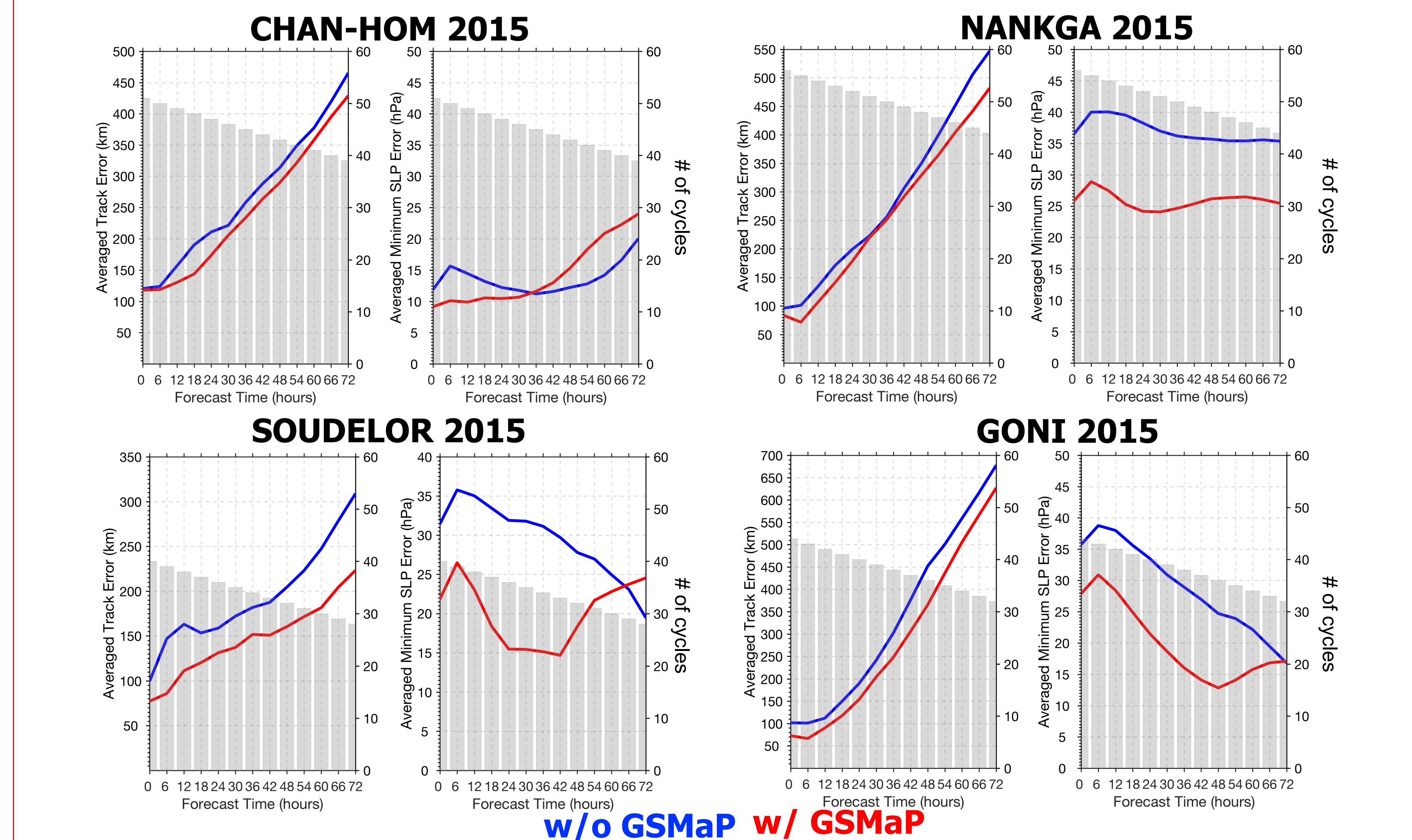
Improved Hydrometeor Analysis



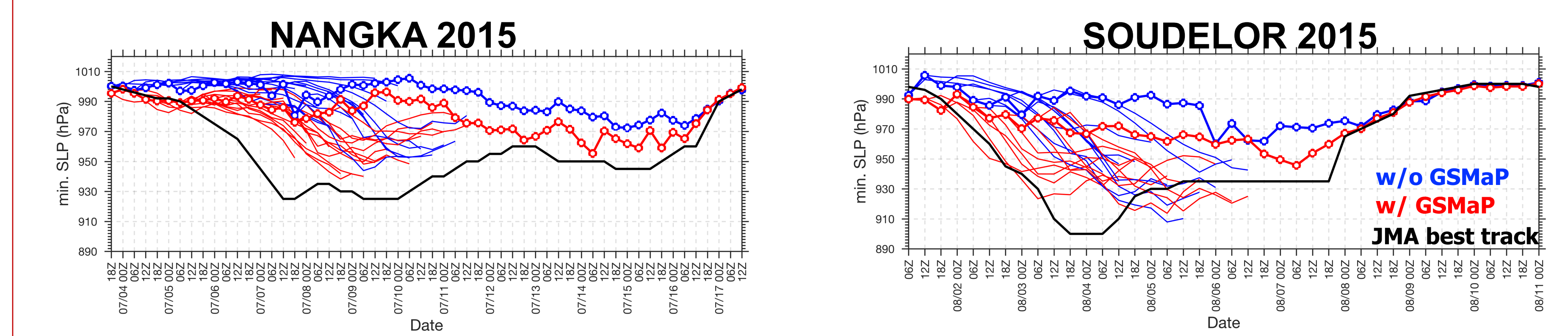
Improved short-range Precipitation Forecast



Improved 3-day Track & Intensity Forecasts



3-day Forecast Initialized at Different Time



Acknowledgements

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